

An investigation into trapezoid packing



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Grouping Problems

- Partition a set of items I into an exhaustive set of mutually exclusive subsets (groups) $U = \{U_1, \dots, U_{|U|}\}$ such that $|U|$ is minimised and

$$\forall U_i \in U, U_i \in F$$

where F is the set of all feasible groups.

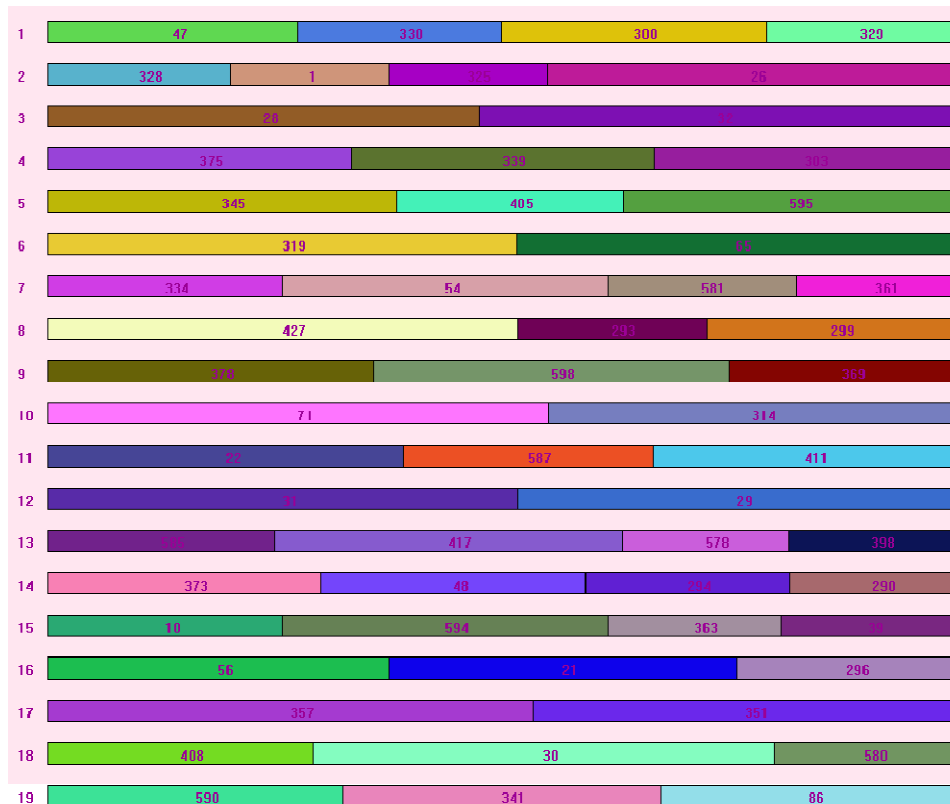
- For example, in one dimensional bin packing and stock cutting, given a “size” s_j for each item $j \in I$,

$$U_i \in F \Leftrightarrow \sum_{j \in U_i} s_j \leq C$$

for some predefined constant C



Grouping Problems



Example 1-D Bin Packing Solution
(most items are different)

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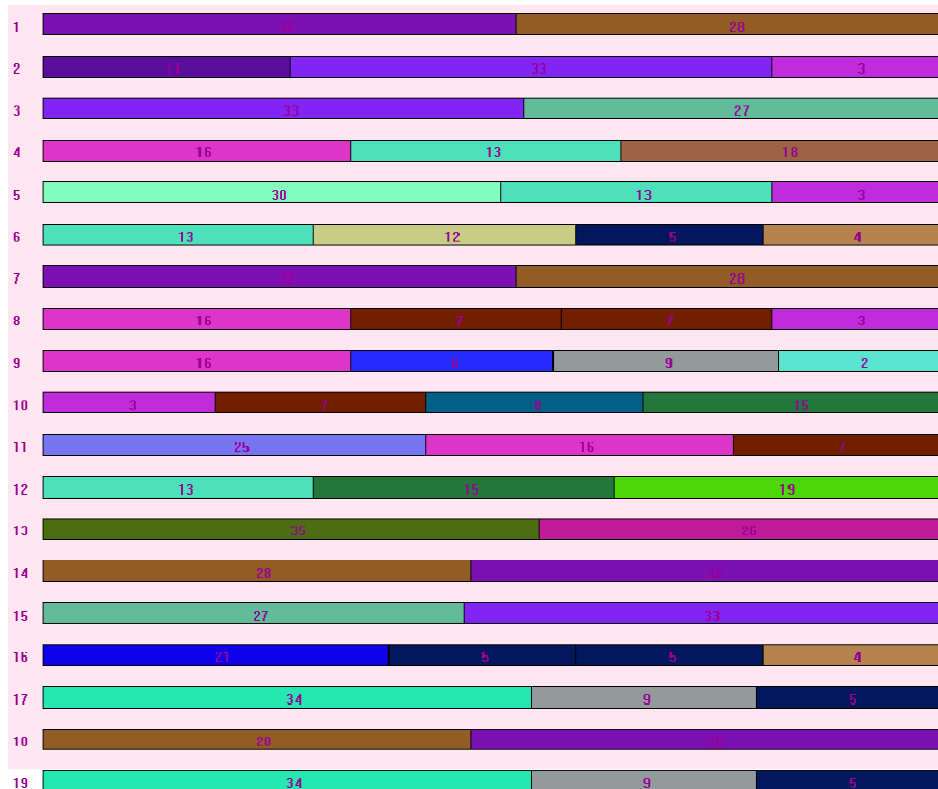
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Grouping Problems



Example 1-D Stock Cutting Solution
(many items of same dimensions)

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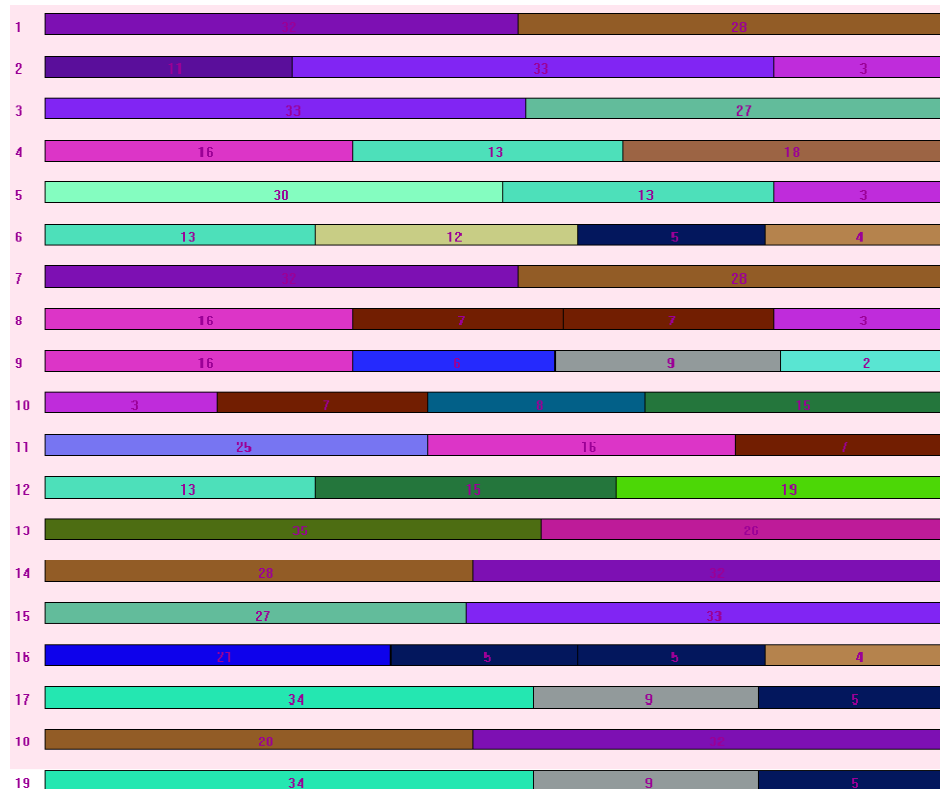
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Grouping Problems



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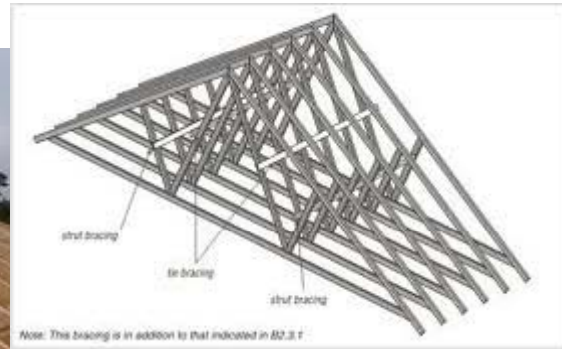
■ In these cases group ordering unimportant.

■ Also, the order and orientation of items *within* each group irrelevant

■ But what if it is?

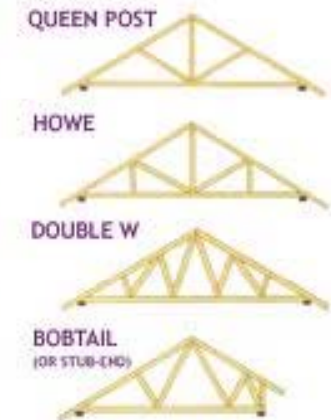
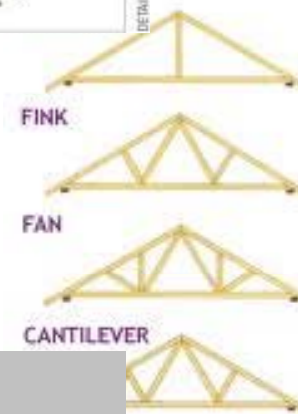
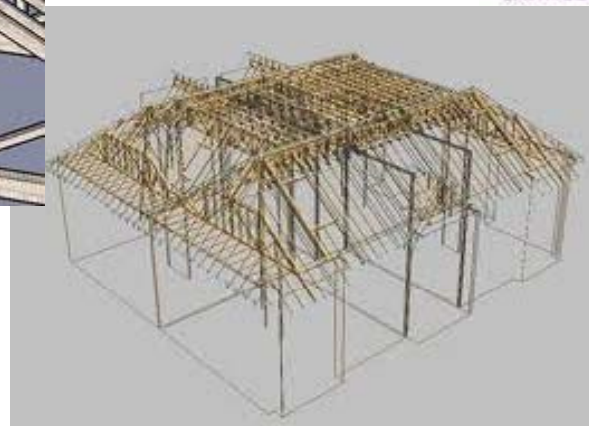
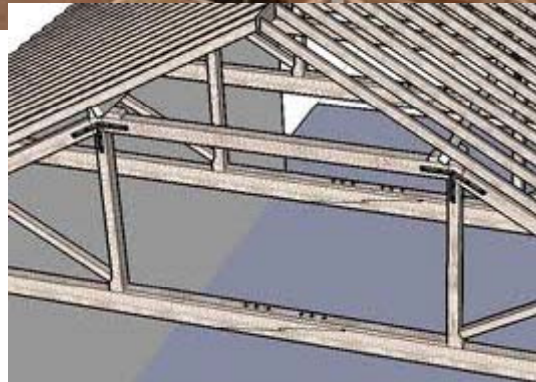


Truss Cutting Problem



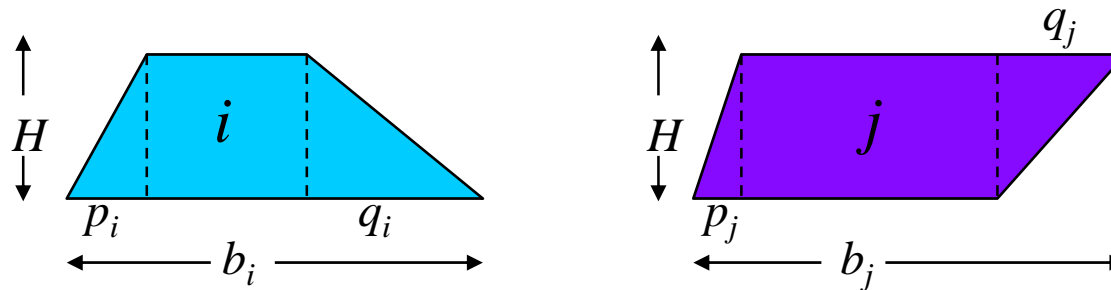
Note: This bracing is in addition to that indicated in B2.3.1

DETAIL B 2.3.2 Prefabricated Roof Trusses - additional bracing

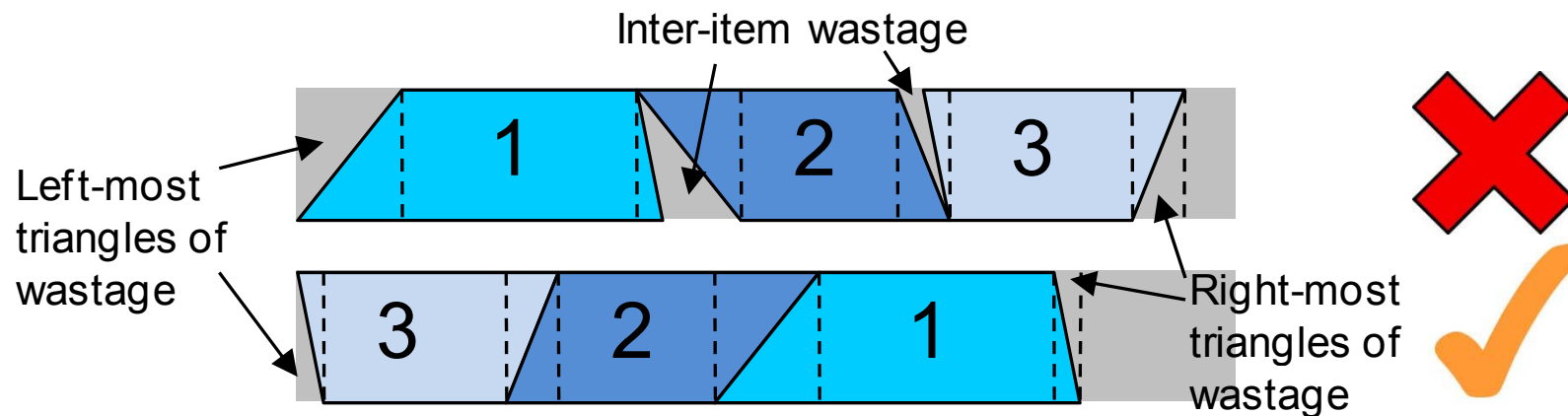




Truss Cutting Problem



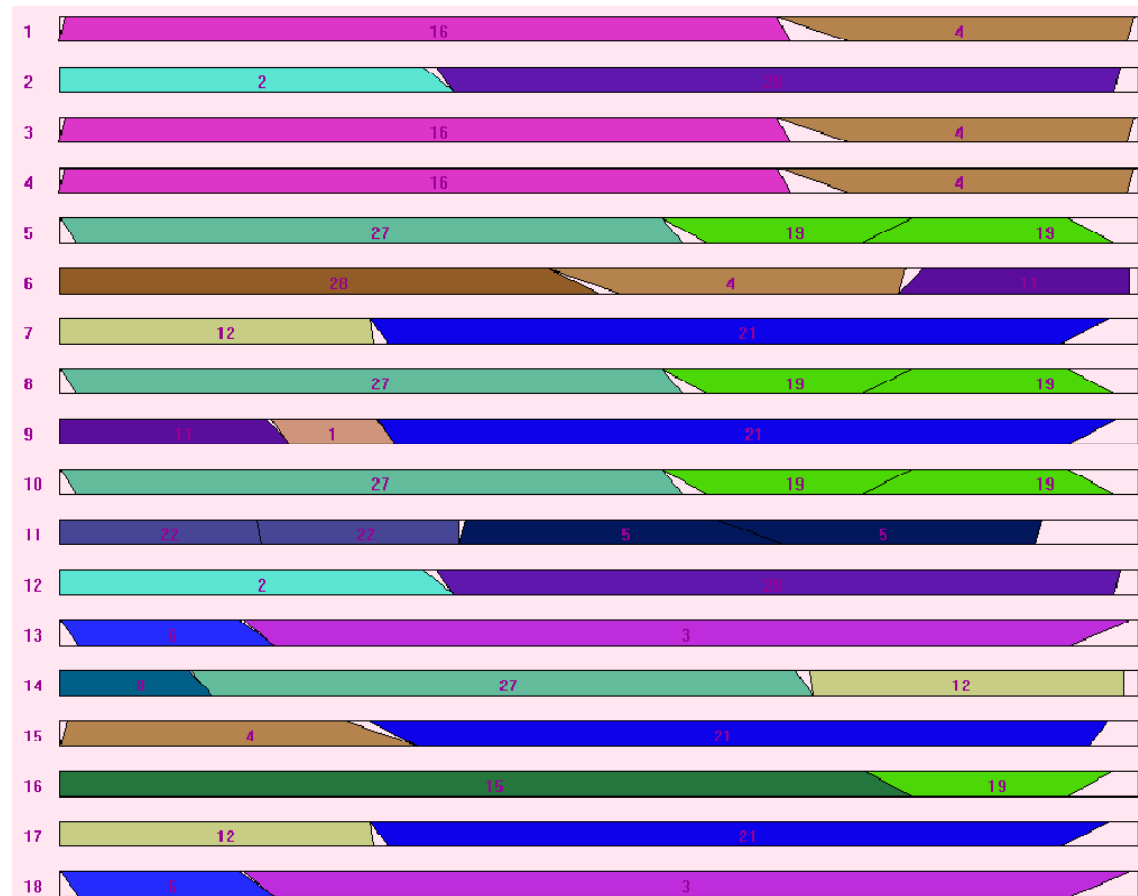
- Given a set of n trapezia of height H , assign each to one of a set of $(W \times H)$ stocks such that the number of stocks is minimised
- Order and orientation of items could be **very important** here...



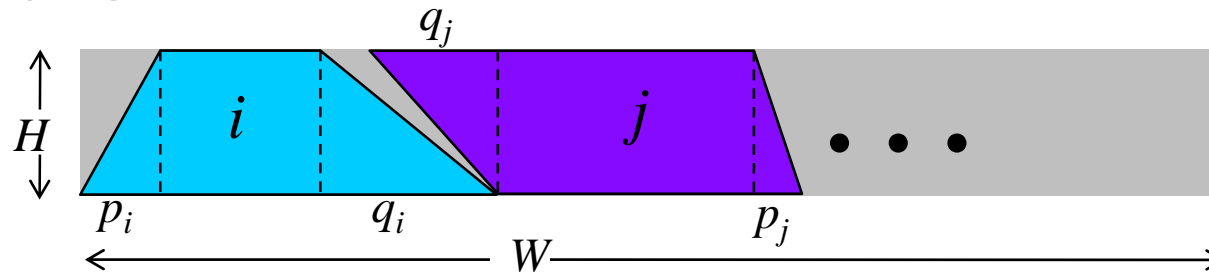


Example Solution...

- There are two issues:
 - Which items should be grouped together? (the **grouping problem**)
 - Given a group of items, how should these be arranged on a stock? (The **truss sub-problem**)



Understanding the Truss Sub-problem



- Items have 4 orientations, but we only need to consider 2
- Inter-item wastage in the above is simply $|q_i - q_j|$
- Wastage w = all inter-item wastage plus the LHS and RHS
- **Problem Def:** Given a group of items U_i such that:

$$A(U_i) = \sum_{j \in U_i} A(j) \leq HW$$

is $U_i \in F$? That is, can we determine an item arrangement such that:

$$A(U_i) + w \leq HW$$

Understanding the Truss Sub- problem

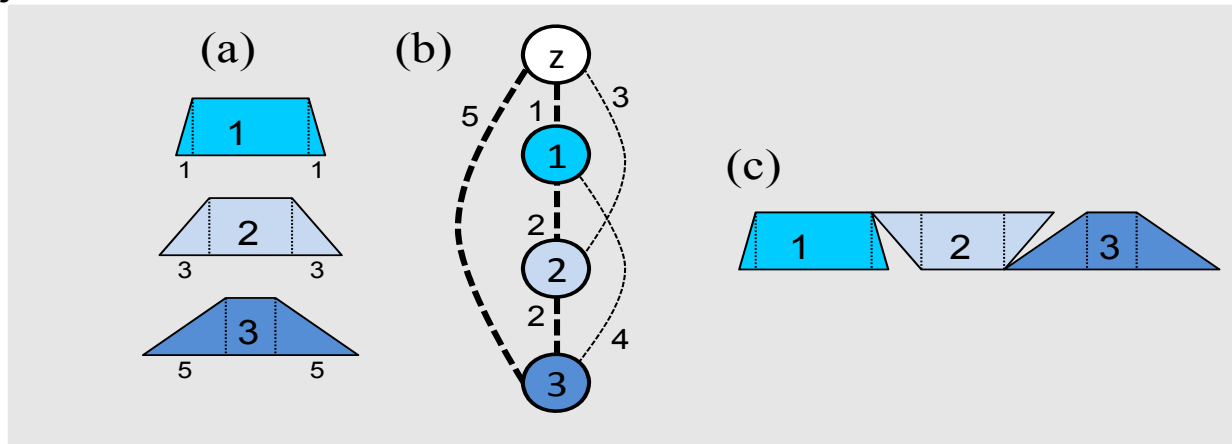


- The Truss sub-problem is a type of TSP
 - Each projection is a “city” ; arcs between cities correspond to wastage; arcs on same shape are set to $-\infty$
 - Two dummy cities (p_z, q_z) represent wastage on the LHS and RHS
 - Can a “valid” route of less than w be achieved (discounting $-\infty$ arcs)?
 - There are $(2^{|U_i|}/|U_i|!)/2$ possible valid routes in total (24 in e.g. above)

Special Cases of the Truss Sub-Problem



- If all projections are of size zero, all items will be rectangular and thus all arrangements are optimal (trivial)
- If all shapes are symmetrical (isosceles trapezia and parallelograms), optimal arrangements can be found by placing items onto the stock in projection-size order...



- However, the general sub-problem is more tricky. Should it be addressed using heuristics, or should we try to use exact methods?



Experimental Set-up

- Two Algorithms were implemented:
 - **A)** Attempt to “solve” sub-problems using simple greedy heuristics. If no suitable arrangement is found, then assume $U_i \notin F$
 - **B)** As above but also employ branch-and-bound for increased accuracy (**Xpress**). During a run, save information on previous B&B applications to avoid duplication of effort.
- Both embedded in a local search algorithm originally designed for 1D bin packing (Lewis, R. (2009), *Comp. and Op. Res.*, vol. 36(7), pp 2295-2310).
- Over 1000 problem instances were generated attempting to simulate the “real” problem (not totally satisfactory...)



Example Solutions

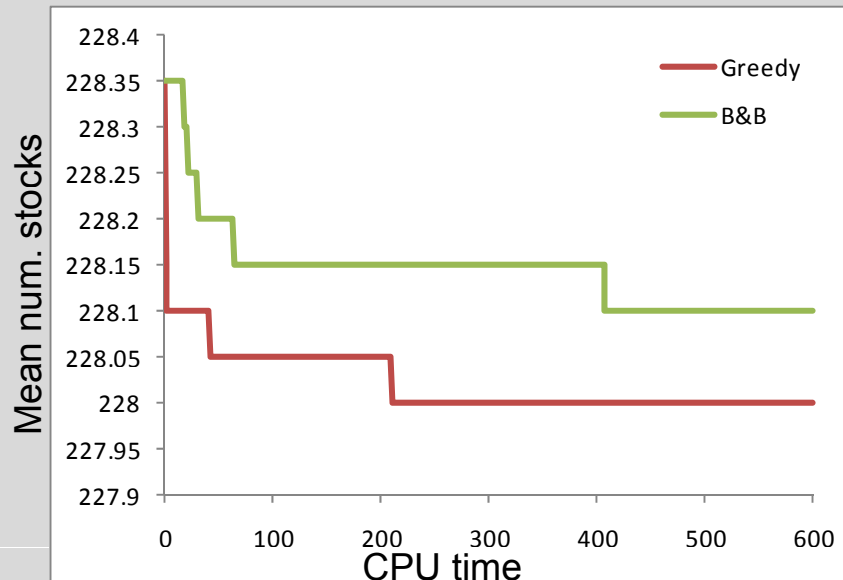
- Sub-problems in the simulated “real” problems typically very small...
 - With $n = 500$, theoretical minimum never reached, but answers are within other generated upper and lower bounds

- We also produced further problems with larger sub-problems...
 - With $n = 500$, theoretical minimum reached in over half of cases

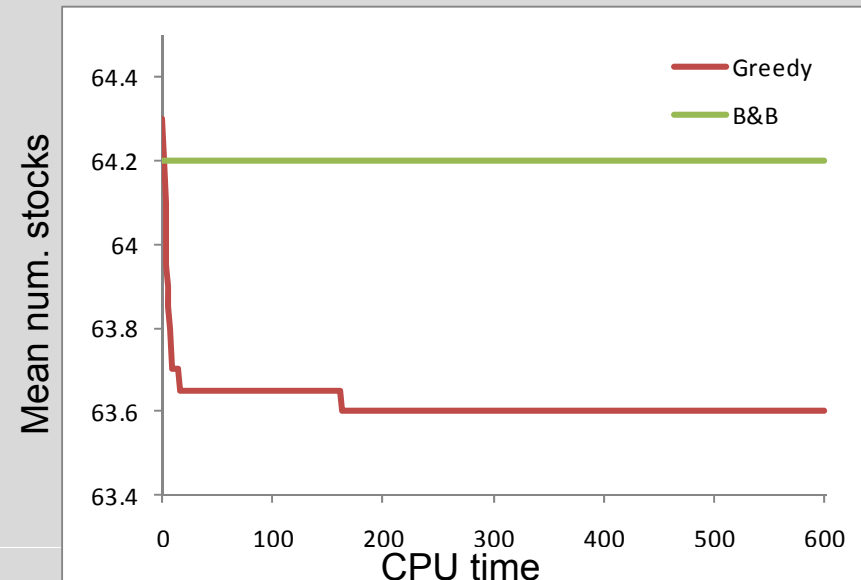
Run Profiles (no repeated items)



Small Sub-problems (approx 2.5 items per stock)



Big Sub-problems (approx 9 items per stock)

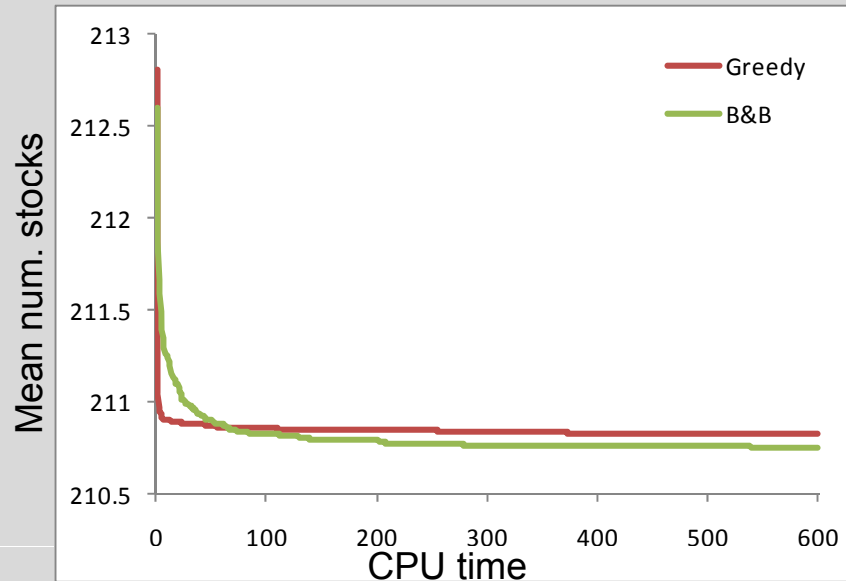


- Algorithm using branch and bound is much slower
- This is accentuated in problems with smaller trusses (larger sub-problems)
- Despite its increased accuracy, the cost of B&B means less iterations of the local search algorithm take place over time

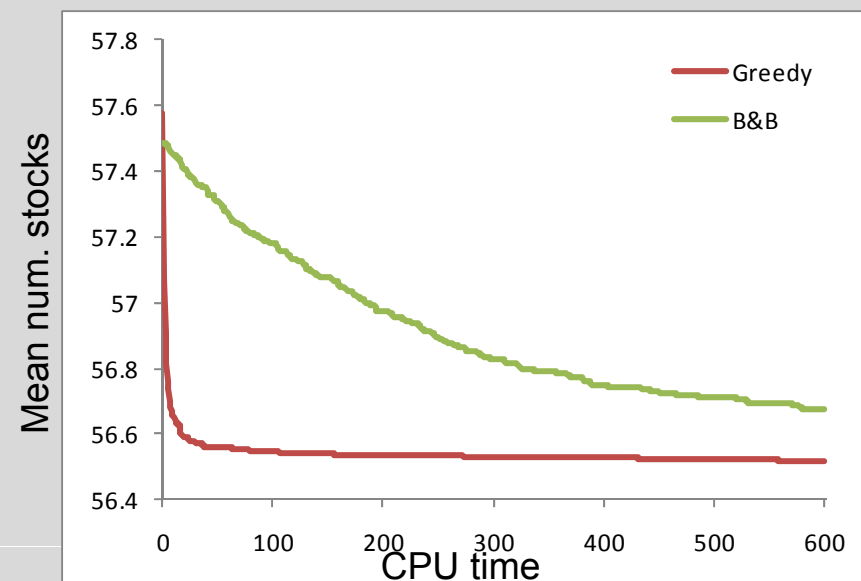


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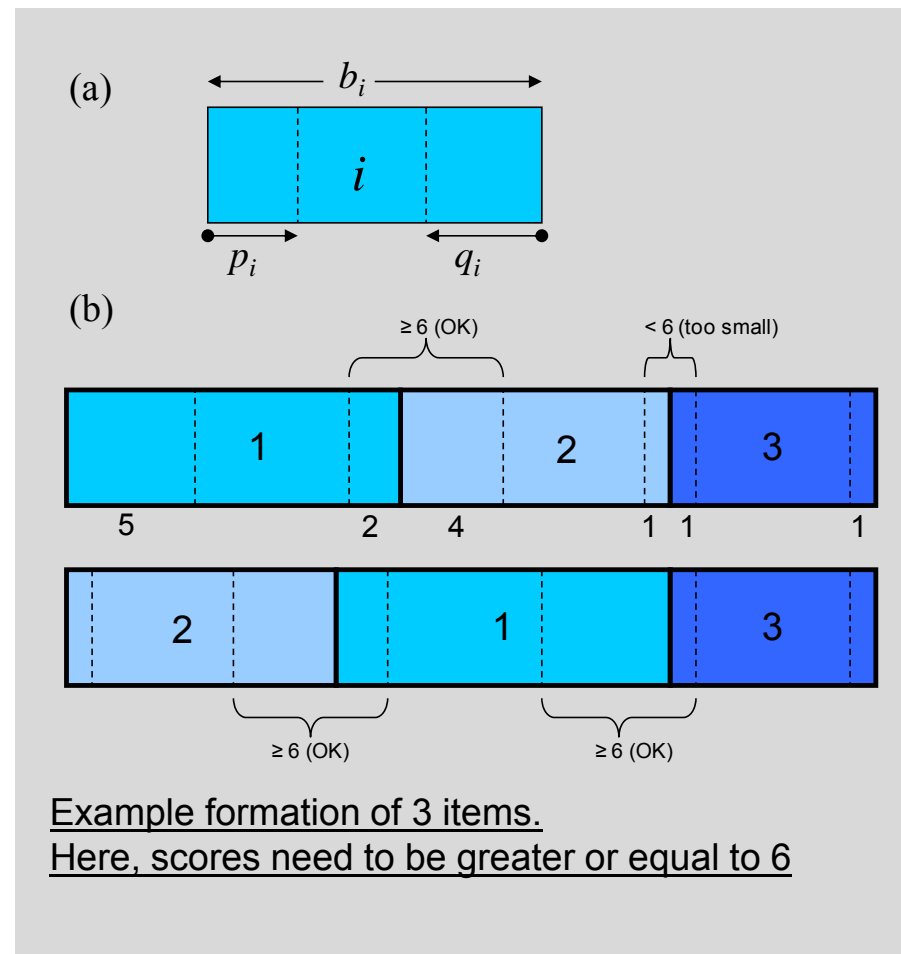
- ▶ However, performance of B&B algorithm improves when items are repeated in the original problem (the same sub-problems are encountered more often)
- ▶ Though larger sub-problems still lead to degradation of performance

A Related Problem

(Goulimis (2004) “Minimum Score Separation – an open combinatorial optimisation problem” *J. of Op. Res. Soc.*, vol. 55, pp 1367–1368)



- In the manufacture of cardboard boxes, we need to cut rectangular strips from stocks which are also simultaneously scored.
- Each item has two “score distances”. However, scores cannot be too close together when cut
- Thus, like the truss problem, item order and orientation is considered when deciding whether $U_i \in F$
- Our truss algorithm is applicable here with minor alterations





Conclusions

- These problems are similar to bin packing and stock cutting problems. But they also involve potentially complex sub-problems
- During optimisation, many sub-problems are faced. Thus an appropriate balance needs to be struck between accuracy and expense in solving these.
- Saving information on the outcome of previous sub-problems has benefits when items are repeated in a problem
- Future Issues:
 - Truss Problem may have multiple sized stocks in practice.
 - Might also be interested in using “left-overs” for solving the next problem
 - Can improved performance be achieved? In particular by improving the methods by which sub-problems are solved.



Conclusions

- These problems are solved by cutting stock initially
- During an approach, accuracy is faced. Thus between
- Saving problem is sub- ed in a
- Future Issues:
 - Truss Problem may have multiple sized stocks in practice.
 - Might also be interested in using “left-overs” for solving the next problem
 - Can improved performance be achieved? In particular by improving the methods by which sub-problems are solved.

Thank you...